



Munich Personal RePEc Archive

Text-mining IMF country reports - an original dataset

Mihalyi, David and Mate, Akos

Central European University

2 August 2019

Online at <https://mpra.ub.uni-muenchen.de/100656/>
MPRA Paper No. 100656, posted 27 May 2020 06:21 UTC

Text-mining IMF country reports - an original dataset

David Mihalyi[†]

Akos Mate[‡]

August 2, 2019

Abstract

This article introduces an original panel dataset based on the text of country reports by the International Monetary Fund. It consists of a total of 5561 Article IV consultation and program review documents, published between 2004 and 2018 on 201 countries. The text of these reports provide indications of the perceived policy weaknesses, economic risks, ongoing reforms and implemented or neglected policy advice. Thus the content of IMF reports are widely used in the economics, political science and IR literature. To our knowledge this is the first comprehensive dataset that aggregates these country reports.

The paper gives a detailed account on the data acquisition and management process. To demonstrate and validate the dataset's application for research we present three validation exercises. We find that Article IV reports can indicate incoming institutional reforms, show changes in IMF policy advice over time and identify potential gains from recently discovered natural resources in certain cases. Taken together, this paper contributes an original dataset of IMF country reports and demonstrates how it can be a useful foundation for further research into the role of international financial institutions.

Keywords: *economic policy, IMF, text analysis, original dataset*

JEL classification: E60, F53

*We would like to thank Cristina Corduneanu-Huci, Michael Dorsch, Evelyne Hübscher, Kai Gehring, Saliha Metinsoy, Bernhard Reinsberg and participants of CSAE 2019 and PEIO 2019 conferences for their valuable comments and David Deritei for the help with the python code implementation.

[†]Corresponding author. Central European University and Natural Resource Governance Institute. Email: davidmihalyi@gmail.com. Phone: +36304376393

[‡]Central European University. Email: mate_akos@phd.ceu.edu

1 Introduction

In this paper we are presenting an original panel dataset which contains the corpus of the country reports published by the International Monetary Fund (IMF) between 2004 and 2018 for 201 countries.¹

The IMF country reports are one of the main go-to sources for economists and social scientists to gain insight on the latest economic developments and discussions of policy reforms under way. These reports prepared by IMF staff roughly once year provide a unique insight on their 189 member states. In the past, many cross-country studies have built on the content of these country reports, analyzing the policy advice they provide (Broome, 2015; Gallagher and Tian, 2017; Ortiz et al., 2015; Rodrik, 2006; Roy and Almeida Ramos, 2012), the conditionalities linked to the loans (Kentikelenis et al., 2016), (Mussa and Savastano, 1999) and the critical assumptions underpinning the analysis (Blanchard and Leigh, 2013). Research by Shin and Glennerster (2003) found that countries face lower borrowing costs when they opt to make the content of their IMF reports public. In addition to the economics literature, the IMF country reports are also being used in the political science and international relations fields as well. Lombardi and Woods (2008) looks at the various outputs of the IMF’s (including country reports) through an IR theory lens and examines whether they promote learning and socialization. Using data from IMF’s Monitoring of Fund Arrangements database (Dreher et al., 2015) analyzes the connection between IMF conditionalities and a country’s political importance. This non-exhaustive list of research demonstrates that IMF reports serve as an important data source. With regards to their research method, the works cited above rely, at least partially, on the qualitative review of country reports. For example, (Broome, 2015) reviews all country reports for four countries over multiple decades, (Gallagher and Tian, 2017) reviews 528 reports for 33 countries in 16 years and (Ortiz et al., 2015) reviews 616 reports globally published in a 5 year window.

With the advances in computer assisted text analysis it became possible to quantitatively assess large bodies of text, which previously would have required vast amount of hand-coding. It was shown that quantitative content analysis is a viable (and often better) technique when compared to qualitative coding by experts (Laver and Garry, 2000; Laver et al., 2003). Political science research has been experimenting with quantitative content analysis for some years now and developing novel methods to exploit the huge amount of text data available (Lowe, 2008; Grimmer and Stewart, 2013). The economics literature similarly started to make use of texts as data, which is surveyed in a recent paper, where Gentzkow et al. (2017) reviews the possible techniques and use cases for economic analysis.

Applications include using central bank communications to predict changes in policy rates (Apel and Grimaldi, 2012), fluctuations in Treasury securities (Lucca and Trebbi, 2009) and identifying home bias by analyzing the tone of the speeches of the members of the Governing Council of the Eurozone (Bennani and Neuenkirch, 2017). Similar approaches were used to forecast trends in unemployment by examining Google search queries (Choi and Varian, 2009). Other approaches used newspaper articles to measure policy uncertainty in the US (Baker et al., 2016) and forecast stock prices using the sentiment of newspaper articles relating to particular companies (Tetlock, 2007). Finally, Gehring and Lang (2018) used the tone of credit

¹The dataset is currently available upon request. We plan to release the full dataset with accompanying codes and a codebook upon publication of this companion paper.

rating agency statements to evaluate the impact of IMF programs.

Three papers have deployed automated searches on IMF reports. (Sands et al., 2016) looks at mentions of pandemics before and after outbreaks, (Kenny and O'Donnell, Kenny and O'Donnell) looks at mentions of gender over time and (Beaudry and Willems, 2018) retrieves the names of mission chiefs and looks at their influence on IMF forecasts.

As this brief overview shows, the explosion of technical and methodological advances gave way to a wide range of research applications that provide important insights for social scientists. With our novel original dataset we aim to contribute to this growing body of research by providing a new and exciting way to look at the possible impact of the IMF's country reports. The paper is structured into four main sections. In the first section we briefly cover the context of the IMF's country reports and why they are important data for research. The second section introduces the dataset. It provides details on the methodology of the data acquisition and processing. We also cover the basic descriptive qualities of the panel there and some discussion on the limitations and missing data. The third section provides some cursory glance at possible use cases for the data, such as using a dictionary to look up word frequencies of keywords of interest and associations between such frequencies and policy actions. Finally, in the fourth section we conclude our paper and discuss further avenues for refining the dataset and using it for research.

As Gentzkow et al. (2017, 50) note in their review in the quantitative text analytics literature "virtually all of the methods applied to date, including those we would label as sophisticated or on the frontier, are based on fitting predictive models to simple counts of text features". This is the method we follow in the illustrative examples we present in the validation section.

2 IMF country reports and their contents

The IMF is one of the most influential international financial institution. It engages in monitoring economic and financial policies, offers technical assistance on economic affairs, and provides loans to countries in need. The monitoring of country policies is carried out as part of the consultations based on the Article IV of its Articles of Agreement. In the case of countries receiving IMF financial assistance, additional monitoring takes place through regular program reviews.

IMF country reports are drafted by IMF teams. A small team of IMF economists visits the country in-person (the "IMF mission") to gather data, information and hold discussions with mainly government and central bank officials, but also sometimes private investors, labor representatives, members of parliament, and civil society organizations. Upon its return to headquarters, the staff prepares a report, which forms the basis for discussion by the Executive Board. The Board's views are subsequently summarized and transmitted to the country's authorities. The views expressed in these report are those of the IMF staff team. The views of the Executive Board are summarized in a Public Information Notice (PIN), more recently which is attached to the Article IV report. Comments by the authorities on the staff report are also attached, if any were submitted at the time of the Executive Board discussion. The policy for publication of Article IV staff

reports allows for the deletion of market sensitive information.

In principle, Article IV consultations with members takes place annually. The Fund may decide to place a member on an “extended consultation cycle” that is longer than 12 months but not longer than 24 months. This can be done only if the member does not meet any of the following criteria: the member is of systemic or regional importance; the member is perceived to be at macroeconomic risk; the member is facing pressing policy issues of broad interest to the Fund membership; the member has large outstanding credit to the Fund². Countries under IMF program may also be placed on a 24-month consultation cycle, but will generally have more frequent (semi-annual or quarterly) program review reports, which combine a backward-looking assessment with a forward-looking perspective³.

On April 5, 1999, the IMF Executive Board agreed to a pilot project for the voluntary release of Article IV staff reports. Since February 2004, reports are made public by default unless the country blocks publication. Currently, nine out of ten member countries agree to publication of a Press Release, which summarizes the staff’s and the Board’s views, and four out of five countries agree to publication of the staff report itself. The availability of country report is even higher for program reviews, 96 percent of them are made public.⁴

These country reports follow a similar structure. First, the PIN or Press Release, followed by the Staff Report, then Information Annexes, then additional analysis, then comments by the authorities (if any). The PIN is rather short (approx 3 pages) and summarizes the staff report and describes in brief the IMF Executive Board’s views. The staff report follows the following structure: first it provides some country background, then describes recent policy developments, then outlook and risks, then key policy areas are discussed in-turn: fiscal policy, monetary policy, financial sector, structural and competitiveness policies. A staff appraisal summarizes findings and provides policy recommendations. Data tables conclude the staff report, which provide actual and forecast values of key economic indicators. Additional annexes describe further analysis prepared by the IMF staff, for example debt sustainability analysis or macro-prudential analysis. Finally, authorities may chose to provide comments on the IMF report, though in practice they more often don’t. While the report’s main purpose is to surface any risk to domestic and global stability, in practice these reports touch on a variety of policies deemed economically significant.

3 The IMF country reports dataset

3.1 Scraping, cleaning and constructing the database

We scraped 6347 PDF documents from from the IMF website which were tagged as country reports published between 1st January 2004 and 31st December 2018.⁵ We then stored each document alongside the corresponding meta data displayed on the IMF website: the title of the document, its publication date, a

²Selected Decisions and Selected Documents of the IMF, Issue 39 - Article IV Consultation Cycles. As updated as of March 31, 2017

³Factsheet. As updated as of March 6, 2018

⁴Transparency at the IMF: <http://www.imf.org/en/About/Factsheets/Sheets/2016/07/27/15/35/Transparency-at-the-IMF>

⁵<http://www.imf.org/en/publications/search?when=After&series=IMF+Staff+Country+Reports>

series ID, the URL and the filename. This was done using Webscraper.io ⁶. We then converted these PDF documents into plain text using the PDFtotext tool⁷ and textract ⁸.

We first divided these documents into two groups based on their title: general and thematic country reports. We kept only the general staff reports: either those labelled Article IV Consultations, IMF program reviews or Post-Program Monitoring. This means that we dropped all thematic country reports, including “ROSC” reports on the compliance with various international standards and codes, Financial Sector Assessments, Poverty Reduction Strategy Papers, and Selected Issues Reports which as its name indicates will focus on only a handful of policy areas.

The reason we dropped these thematic reports is because unlike country reports, they don’t conform to similar structure and depth of analysis. The content of a single such thematic report (e.g. 50 pages on Value Added Taxes) may skew the overall body of text for very strongly in one direction. It is also difficult to reconcile how the topics of these reports are selected and their scope defined. While encouraged, publication of thematic reports is voluntary, and there is more variance on whether and when (often years later) they are published. Therefore, unlike general staff reports, the body of text from thematic reports is unlikely to be balanced across major topics of relevance from the perspective of macroeconomic risks. We dropped such thematic reports and were left with 5561 general reports .

We also note that there are 160 reports that relate not to a single country rather to a country group, such as a currency or trade union.⁹ The IMF also writes regular country reports on the individual member states within these groups, and we expect those provide more directly relevant information on the country in question. Therefore we do not use the country group reports in country level analysis, but do include them in the dataset.

The final dataset includes 5561 reports, from 2004 to 2018. We chose 2004 as our starting year, because this is the year when reports became published by default. There are much fewer country reports from earlier periods, and even among these the majority are scanned PDF which make text recognition difficult and imprecise. The make-up of the reports is the following: 160 Country group reports and 5401 individual country reports. We classified the individual country reports into four categories based on their content: (i) Article IV reports, (ii) IMF program documents¹⁰, and (iii) other staff reports. Table 1 shows the distribution of documents.

As the IMF does not provide a consistent categorization for all documents uploaded to their website (the tags available on the website are less reliable especially for older documents) we created our own categorization using the title of the reports. Similarly, the subject year for the Article IV reports were taken from the document title, as sometimes the subject year and publication year can differ, depending when the document

⁶<http://webscraper.io/>

⁷ <https://github.com/jalan/PDFtotext>

⁸<https://textract.readthedocs.io/>

⁹Such country groups are the Baltic cluster, Central African Economic and Monetary Community, Central and Eastern Europe, Eastern Caribbean Currency Union, Euro Area, West African Economic and Monetary Union, and Multi-Country Reports.

¹⁰These include ongoing program evaluation reports, request for certain IMF programs from countries, and various IMF assistance program reports for emergency lending, stand-by agreement, flexible credit programs, debt relief programs and ex-post program evaluation documents.

Table 1: Distributions of documents in the dataset

| Type of document | Count | mean word count | Std.dev of word count | min | max |
|----------------------|-------|-----------------|-----------------------|-------|--------|
| Article IV | 1639 | 33.40 | 10.30 | 13.10 | 87.90 |
| Country group report | 160 | 32.00 | 24.20 | 2.90 | 194.00 |
| IMF program document | 1138 | 34.30 | 14.30 | 3.10 | 150.70 |
| Other staff report | 2624 | 31.20 | 32.30 | 2.30 | 426.90 |

got uploaded to the IMF’s website.

The variables included in the dataset are the following:

Table 2: Variables in the dataset

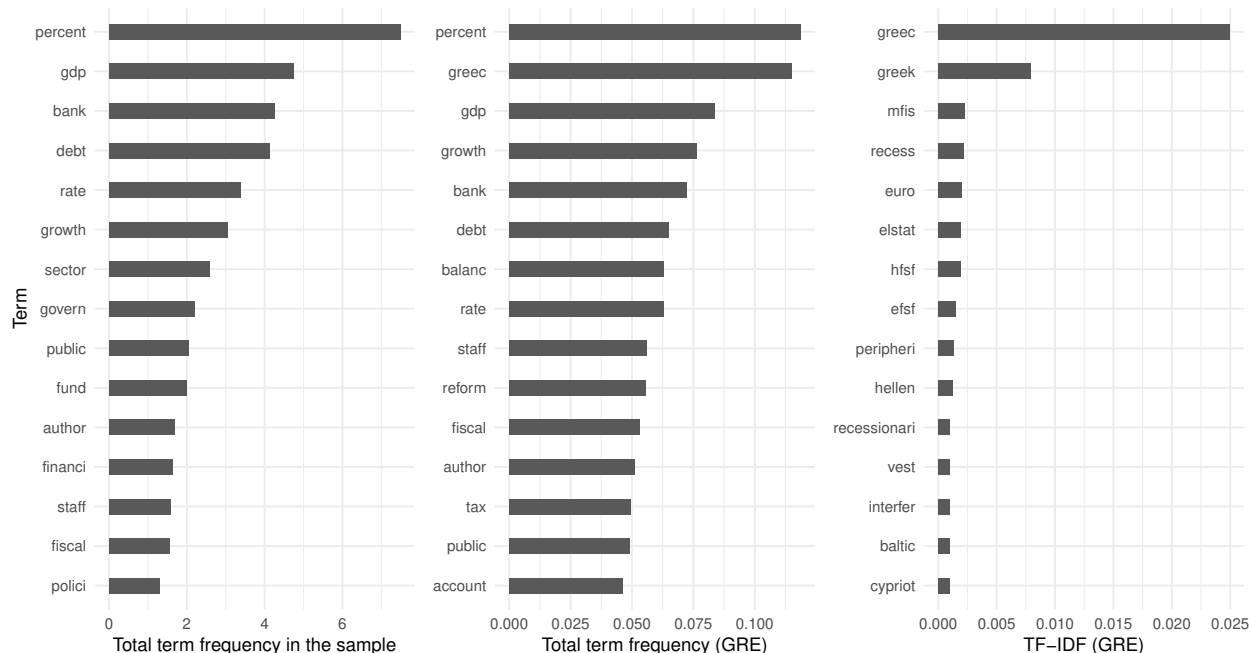
| Variable | Description |
|-----------------|---|
| ccode | ISO 3166-1 alpha-3 country code (Country groups are coded "CG") |
| country | Name of the country being reported (country group reports are coded as "Country group") |
| year_p | Year of the report being published online (present for all documents) |
| year_s | Subject year for Article IV reports (missing for the other type of documents) |
| doc_name | The internal document name of the report (serves as unique ID for the reports) |
| title | Full title of the country report |
| text | The raw plain text content of the report, without any pre-processing or formatting. |
| type | the type of the document. See Table 1 |

3.2 Using the dataset

The dataset allows to search for any keyword across reports and retrieve a clean panel table which summarizes the frequency of mentions of the search term across country and years. The search term can be a single term, such as 'elections', 'protest' or 'default' (unigrams) or a combination of words such as 'raise taxes' or 'labor market reform' (n-grams). The output of the search can be summarized using the following metrics. First, the absolute number of search hits by country year provides a crude metric of relevance, but countries or years with longer or more frequent reports will be over-represented. Second, we can analyze the term frequency (tf), measured for example occurrence in every 1000 word of text. Third, and this is our preferred approach, we can analyze the term frequency-inverse document frequency (tf-idf), which provides a numerical statistic which reflects how important a keyword is within a single document in contrast to its frequency across all other documents. Figure 1 illustrates this point.

The bar chart on the left depicts the highest term frequency unigrams across a random sample of 500 reports and shows that 'GDP', 'percent' and 'bank' occupy the top spots. When looking at a single report (the 2013 Article IV report on Greece), as per the bar chart in the middle, many of the top term-frequency unigrams are the same, though many terms specific to Greece (including the country's name) also appear. The third bar graph shows the term frequency-inverse document frequency for the same Greek document. It shows that words specific to the Greek context of the time show on the top list, including mentions of the recession, words relating to financial sector stability ('MFIs', 'HSFS') and to the Eurozone ('Euro', 'EFSF').

Figure 1: Highest term frequency words in sample and in a single report



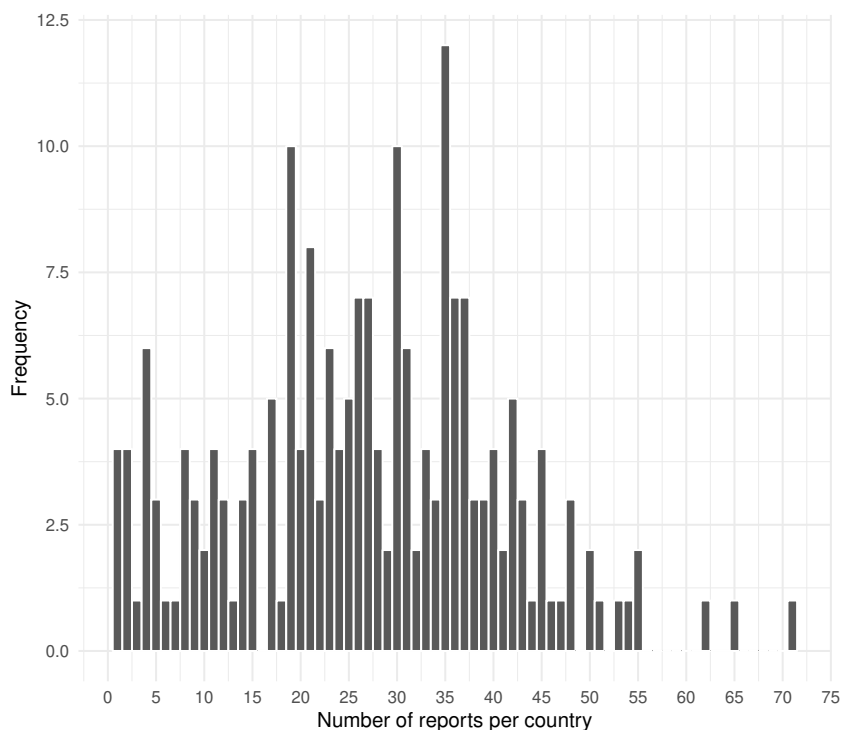
The dataset and accompanying code, which we plan to release publicly alongside documentation allows for

conducting such searches effectively as well as more advanced text analyses.

3.3 Main properties

The panel consists of 5561 observations for the years between 2004 and 2018. The detailed breakdown of country reports per year is presented in Table 3. The distribution of reports per countries is rather uneven (as shown by Figure 2) as a result of some countries receiving more IMF staff visits than others. There are some notable instances of denying access to IMF staff, such as Venezuela and Argentina. The median reports per country is 13, while the mean is 13.5. As for outliers in the report numbers, there are 15 countries that have 5 or less, and 8 countries with 25 or more reports.¹¹

Figure 2: Distribution of country reports per country



In the Appendix, Table 6 gives a more detailed look at number of reports per countries in the panel, as well as the first and last year the country had an IMF report published and average report per year for the period the country participated in the reporting process.¹² However, our dataset covers most of the countries (with some notable exceptions, such as Venezuela). It is also visible that our panel contains more reports for the Central Eastern European, and certain African and Latin American countries.¹³

¹¹Countries with 5 or less reports: Anguilla, Argentina, Bhutan, Brazil, Ecuador, Guyana, Macau SAR China, Montserrat, Nauru, Serbia and Montenegro, Somalia, South Sudan, Syria, Tuvalu, Uzbekistan. Countries with 25 or more reports: Armenia, Ireland, Jamaica, Mexico, Pakistan, Romania, Rwanda, Uganda

¹²This table includes all types of reports in the dataset.

¹³With high report count for Ireland and Greece as well, due to the global financial crisis.

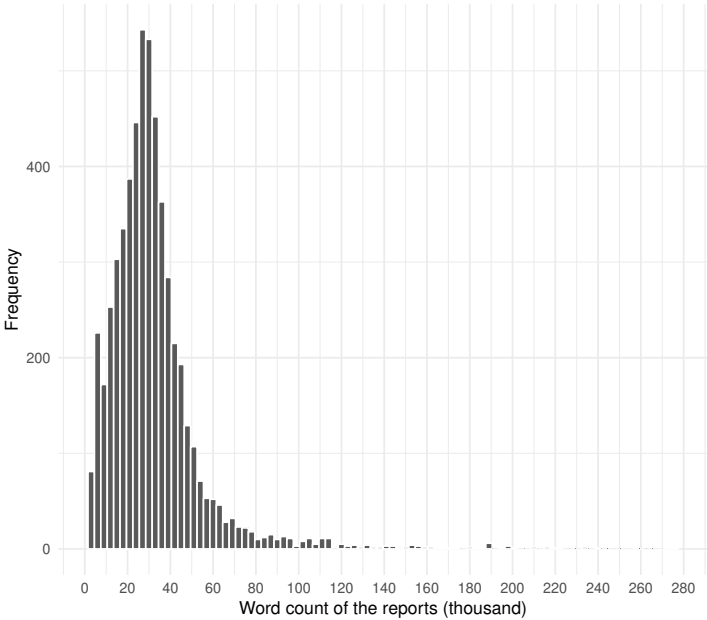
The descriptive statistics for the corpus in Table 3 shows that the mean word count and the standard deviation of the word counts are stable over the panel. However, the minimum and maximum word counts for some country reports display considerable fluctuations over the years.

Table 3: Descriptive statistics of the dataset, grouped by years

| Year | Reports total | Article IV | IMF Program doc. | Mean word count | Std.dev word count | min | max |
|------|---------------|------------|------------------|-----------------|--------------------|-------|--------|
| 2004 | 114 | 68 | 46 | 34.50 | 11.70 | 10.50 | 78.10 |
| 2005 | 174 | 108 | 66 | 34.10 | 12.10 | 3.70 | 87.30 |
| 2006 | 191 | 113 | 78 | 32.70 | 11.50 | 3.10 | 64.80 |
| 2007 | 175 | 110 | 65 | 30.90 | 9.60 | 4.20 | 60.50 |
| 2008 | 181 | 106 | 75 | 32.30 | 15.20 | 5.30 | 150.70 |
| 2009 | 206 | 107 | 99 | 31.40 | 11.90 | 4.80 | 95.00 |
| 2010 | 226 | 122 | 104 | 32.40 | 11.10 | 4.30 | 70.70 |
| 2011 | 214 | 108 | 106 | 31.30 | 11.90 | 4.30 | 82.90 |
| 2012 | 184 | 103 | 81 | 33.90 | 12.30 | 12.70 | 111.80 |
| 2013 | 190 | 112 | 78 | 35.20 | 15.00 | 10.90 | 125.70 |
| 2014 | 195 | 114 | 81 | 33.70 | 12.20 | 13.60 | 122.80 |
| 2015 | 189 | 113 | 76 | 34.20 | 10.40 | 5.50 | 78.10 |
| 2016 | 179 | 118 | 61 | 37.50 | 10.70 | 17.10 | 75.20 |
| 2017 | 177 | 119 | 58 | 36.50 | 11.50 | 15.90 | 87.90 |
| 2018 | 182 | 118 | 64 | 37.50 | 10.90 | 6.60 | 72.90 |

The distribution of word counts is shown in Figure 3. Due to the verbosity induced by the global financial crisis, the distribution is somewhat skewed. The distribution and descriptive statistics for the sentence count in the corpus tracks the word counts.

Figure 3: Distribution of the word count in the corpus



3.4 Limitations of the dataset

One key issue that researchers should pay attention to is that the availability of country reports is not distributed uniformly in the data set. Not all countries get the same amount of surveillance, where some countries which might be deemed at lesser risk or of smaller global significance may be monitored in less detail given resource constraints of the Fund. On the other hand, countries with higher macroeconomic vulnerabilities or undergoing IMF program may get more in depth monitoring.

Another factor is that in some cases the country authorities may refuse to have the Article IV published. IMF publications on transparency reveal this happened in about 20 percent of cases in 2004-2005, when our dataset starts and gradually declined to 5 percent of cases in 2014-16 ¹⁴. An earlier study by Edwards et al. (2011) finds that more democratic governments are more likely to release reports, as well as a strong variation in regional patterns (most notably less report in Latin America).

These matter for the research design. The likelihood of a key word appearing at all in reports will depend on the likelihood of the report being published and its depth. The frequency of certain themes may also affect whether the report gets approval from authorities to be published.

We build on Edwards et al. (2011) in the selection of explanatory variables and review how the following variables affect report availability and length of reports.

Table 4: Independent variables for the Heckman Two-Stage model

| Variable | Description and source |
|-----------------|--|
| Population | in log form from World Bank |
| GDP per capita | in log form from World Bank |
| Debt service | as percentage export earnings from World Bank |
| IMF program | in place for at least 5 months in the year (Dreher, 2006) updated |
| Polity IV score | on a -10 to 10 scale (polity2) from Quality of Government dataset. |

We analyze the availability and length of reports using a two-step regression (or Heckman correction). First, we look at the likelihood of having a public report in a given year in a given country using probit regressions depending on whether the country is going through an IMF program and its polity IV score. We present marginal coefficient plots for polity score 4 and whether the country has an IMF program that year 5. In the second stage, conditional on having a report published, we look at how the population, income and debt variables affect the average length of reports (thousands of words) by country in using linear regressions. The results of the Heckman Two-Stage model is in Table 5

We find that there is a higher likelihood of having a report in instances where countries are undergoing IMF programs at the time of assessment. An IMF program increases likelihood of coverage in any given year by over 10 percentage point. We also have higher report coverage in country years with higher institutional

¹⁴Source: Key Trends in Implementation of the Fund's Transparency Policy, IMF

Table 5: Heckman two-stage model results

| | <i>Step 1:</i> | <i>Step 2:</i> |
|-------------------------|------------------------|------------------------|
| | Report year | Word count (000) |
| Polity IV score | −0.0022 (0.0060) | |
| IMF program | −0.4416*** (0.1228) | |
| Constant | −0.0330 (0.0405) | |
| log(GDP per capita) | | −3.8067*** (0.3703) |
| log(Population) | | 0.7514*** (0.2238) |
| Debt service | | 0.0110 (0.0299) |
| Constant | | 25.3600*** (9.1131) |
| Observations | 1,355 | |
| R ² | 0.2049 | |
| Adjusted R ² | 0.1999 | |
| ρ | 1.1725 | |
| Inverse Mills Ratio | 32.7064*** (9.3387) | |

Note: *p<0.1; **p<0.05; ***p<0.01

Figure 4: Likelihood of report availability depending on Polity IV score

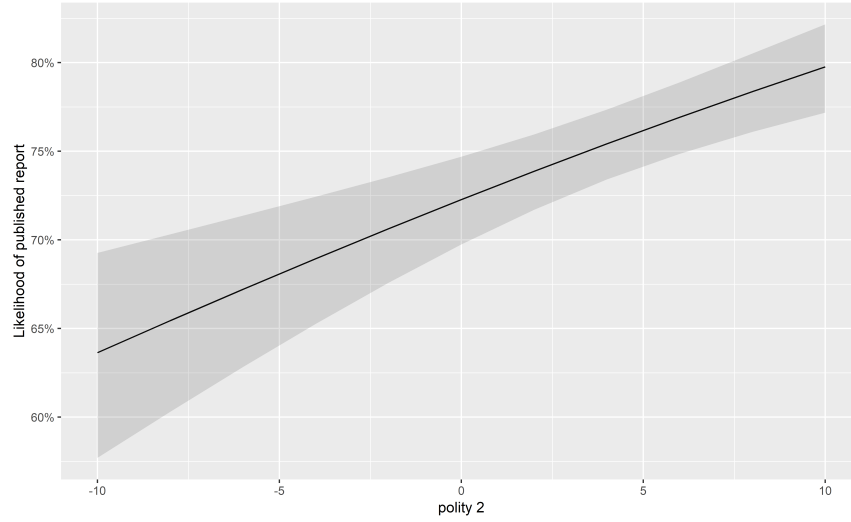
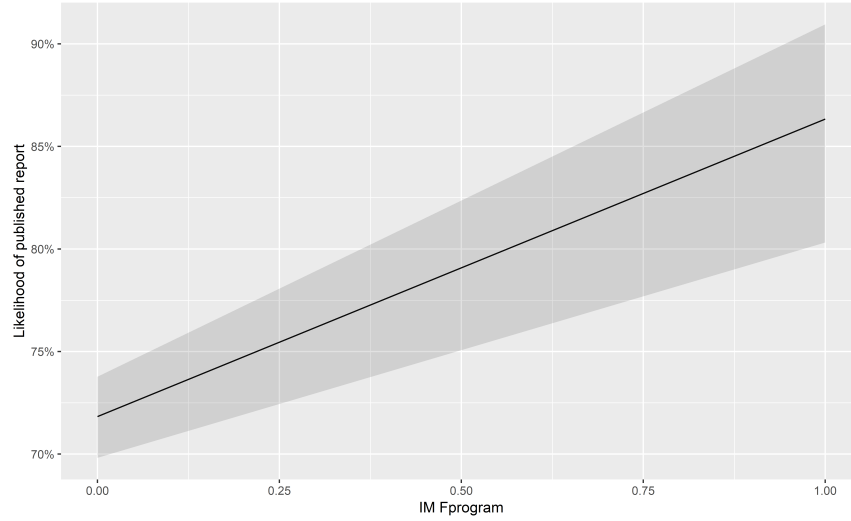


Figure 5: Likelihood of report availability depending on IMF program taking place



scores as measured by polity. Countries with lowest score are 15 percentage point less likely to report than those at high scores.

Conditional on having a report, we find that they are shorter for richer countries. Population and the relative size debt of debt service (a proxy for economic vulnerability) does not seem to make a difference.

Because rare keywords are more likely to appear in countries with more reports and longer text, measures of keyword appearance may be somewhat biased towards larger, less wealthy and more democratic countries and years of financial difficulty. These factors need to be taken into account when designing research using

the dataset. One avenue to address the bias is to measure the relative frequency of appearance of chosen keywords.

4 Demonstration and validation

The below section presents demonstrations at how the dataset may be used for applied in research. These short illustrations are not intended at providing substantive new empirical contributions. Rather they are meant to showcase and validate that text analytic metrics capture meaningful and important characteristics of IMF country diagnosis and advice.

For the analysis we used the R package "quanteda" (Benoit, 2018) to create a document-feature matrix from our corpus, which allows for extracting various word frequency measures.¹⁵ We created a document-feature matrix with single words as tokens and with bi-grams (two word combinations) as tokens.

Our first illustrative case look at the distribution of a keyword (oil) across countries, the second one looks at discussion of reforms vs actual reform events (fiscal rules), and the third illustrates changes in IMF policy priorities over time (consolidation vs stimulus).

We review whether the countries which are considered resource dependent based on conventional economic metrics are the ones where oil is most frequently discussed in country reports. We calculated the frequency of appearance of the word oil across reports, using term frequency with inverse document frequency weighting. We contrast that with fuel exports as percentage of all merchandise export (from World Bank). We chose the fuel exports measure because it is the most widely used measure of oil dependence and has high cross-country coverage. We plotted the results on figure 5.

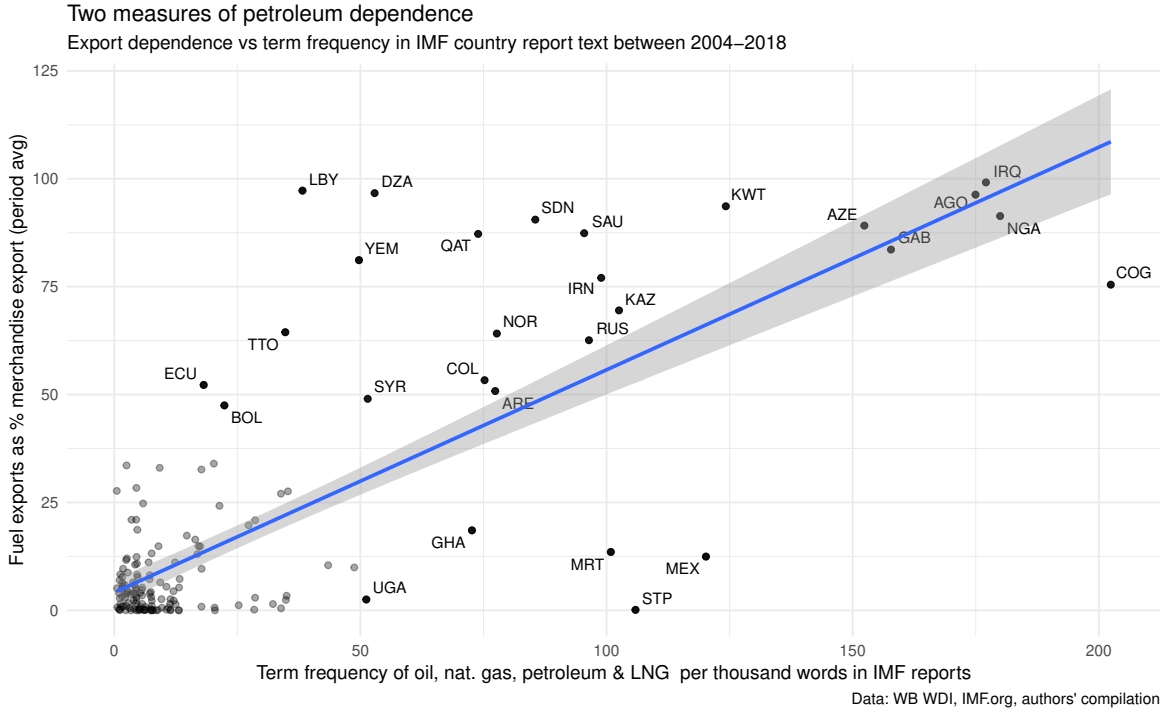
We find a strong positive correlation, with an R-squared of nearly 50 percent. Most countries export no oil, or oil export are tiny in fraction to other goods, and in their reports oil is infrequently mentioned. The 5 countries where oil is most frequently mentioned are in order: Chad, Nigeria, Republic of Congo, Iraq, Gabon, all extremely oil dependent countries (note that no export data for Chad, hence not displayed on plot). This reaffirms, that IMF reports focus heavily on the key sector in these countries.

Another interesting insight comes from the countries below the trend line, where oil is mentioned more than the level of resource dependence would directly imply. Among these are 4 new or prospective oil exporters Uganda, Ghana, Sao Tome, Mauritania, where oil export is low, but IMF reports discuss the expected rapid ramping up of oil production and exports in the future.

This example validates and demonstrates that text search can be used to capture the perceived salience of various macroeconomic risk factors across reports.

¹⁵In some cases in the literature document-term matrix is also used and refers to the same concept.

Figure 6: Two measures of oil dependence



4.1 Comparing actual reform and reform intent

We have conducted bigram search on the term 'fiscal rule' and related terms ('expenditure rule', 'debt rule', 'balance rule', 'deficit rule', 'revenue rule'). This allows to analyze whether fiscal rules are being more frequently discussed in IMF country reports prior to adoption or major changes to existing fiscal rules. The data on the fiscal rules is based on the IMF FAD Fiscal Rule dataset (?).

Fiscal rules are mentioned in 25 percent of all reports. Figure 6 shows that there has been a gradual increase in the number of fiscal rules across countries, and they are also being mentioned in increasing frequency across reports especially since 2008.

Figure 7 shows the frequency of search term appearance depending on whether the country has a fiscal rule, has no fiscal rule, and when the fiscal rule is being implemented or being reformed. Fiscal rules are mentioned at all about twice as frequently in countries which currently have fiscal rules (35 percent of reports), than in countries without fiscal rules (18 percent of reports). They are also being discusses with increasing frequency over time across all groups. But the graph also shows a sharp increase in mentions (and especially term frequency, which takes into account how much it is mentioned in a single report) before first implementation of a fiscal rule and major reform of the fiscal rule (as per IMF FAD dataset). This suggests that IMF is closely monitoring reforms to fiscal rules and probably in many cases providing its own advice.

This example highlights how text search can be used to capture the sequencing of IMF advice and reform

Figure 7: Percentage of countries with fiscal rules and reports with fiscal rule mentions by year

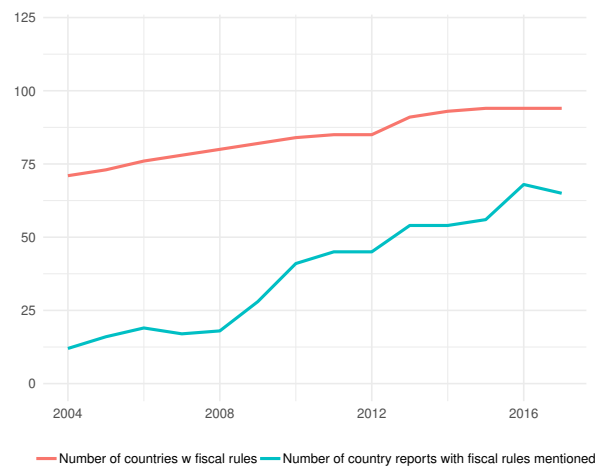
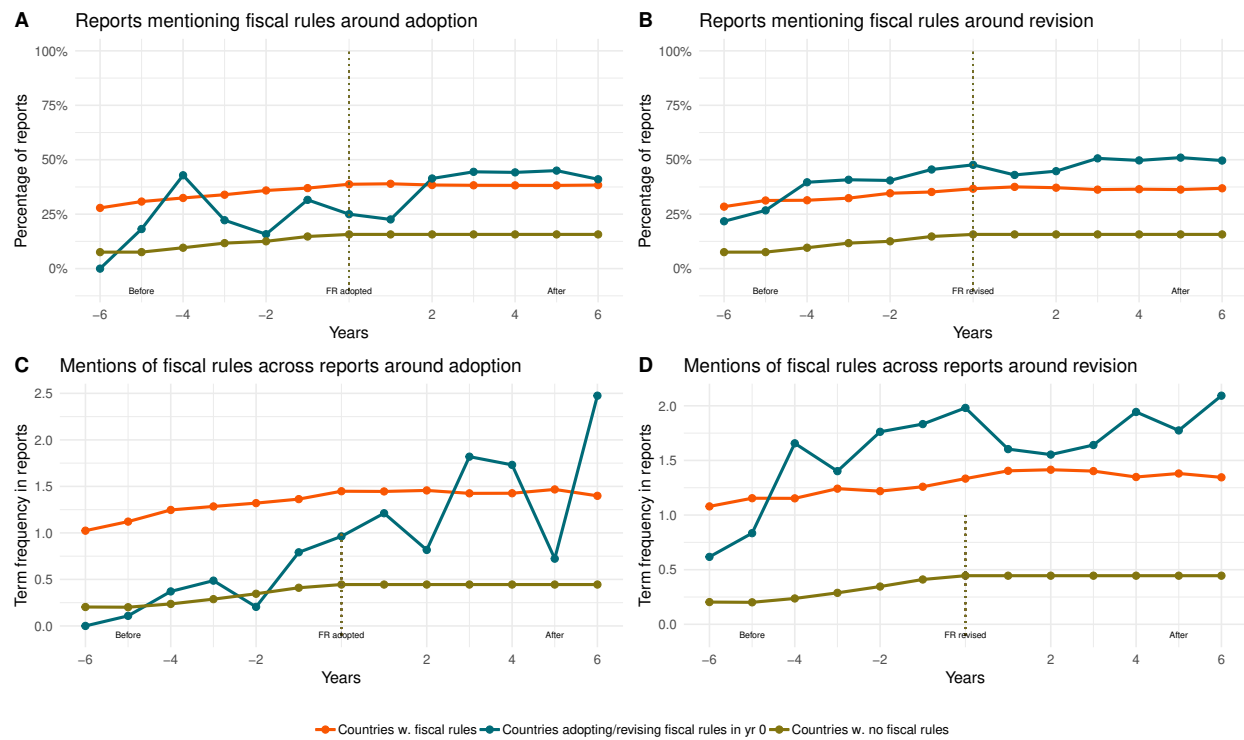


Figure 8: Mentions of fiscal rules vs actual fiscal rules



events. It also validates that such fiscal policy reforms are indeed reflected in our dataset.

4.2 Changes in perceived policy priorities

The dataset also enables to monitor changes to perceived policy priorities. We have constructed dictionaries that describe opposite fiscal policy directions.¹⁶

Table 6: Dictionaries used

| Dictionary | Items |
|----------------------|--|
| Fiscal consolidation | fiscal consolid*, fiscal discipl*, restor* fiscal, fiscal slip*, fiscal solv*, fiscal adjust |
| Fiscal stimulus | fiscal stimul*, stimul* package, fiscal expans* |

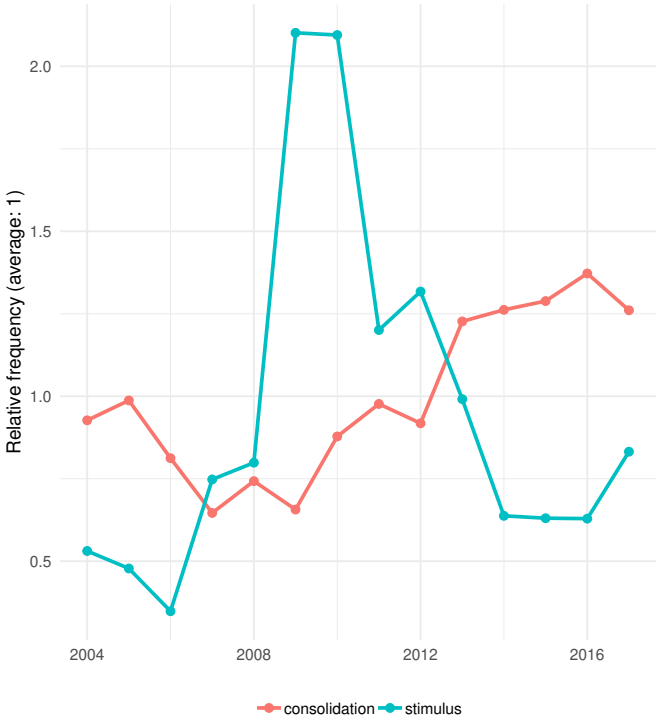
The longer list of key words associated with consolidation appears much more frequently than the words associated with stimulus. But when evaluating relative frequency of their appearance, we can discern trends over time. We find a sharp increase in mentions of stimulus in 2009 and 2010, then followed by a rapid decline. In the meantime fiscal consolidation mentions have increased starting in 2010.

This mirrors the patterns described in (IMF, 2014) and (Dhar, 2014) which discusses IMF policy advice in response to the financial crisis. In fact, the so-called Triennial Surveillance Review, manually reviewed their country reports to map the policy advice it had provided on a sub-sample of 24 countries. They found that across these 24 countries short-term stimulus was recommended in 3/4 of IMF article IVs in 2009 (IMF, 2014).

This illustration provides an example on how text search can be used to map whether changes in global policy pronouncements are being reflected in country reports. The fact that our dataset mirrors the analysis carried out using manual review of the text also validates our approach.

¹⁶The * denotes a wildcard, e.g.: restor* will find *restoring* as well as *restoration*

Figure 9: Relative frequency of mentions of keywords associated with consolidation and stimulus by year



5 Conclusion

IMF country reports are a treasure trove of information on the economic and policy developments of countries across the world. They also provide a window into the policy priorities and advice that the IMF provides. We presented a new dataset which builds on the content of these country reports and show how simple text analytic techniques can be used to gain new insight on this important international organization. This includes studying the length of document and the frequency of mentions of specific keywords over time, across countries and in the years surrounding policy change or economic shocks.

Subsequent refinement of the dataset may decompose reports into its chapters, enrich the meta data with authors and exact dates of drafting. In future work, this dataset can be used to study the factors which may influence the priorities and the overall tone of IMF surveillance across countries and over time. It can be used to examine financial market response to country report findings. It also enables to study if and when IMF policy recommendations are being followed through. Taken altogether, subsequent work using the dataset can help disentangle the determinants and impacts of the IMF's surveillance work.

6 Appendix

Table 7: Number of reports per country in the panel

| Country | ccode | First report | Last report | No. of reports | Mean |
|----------------------|-------|--------------|-------------|----------------|------|
| Aruba | ABW | 2005 | 2018 | 8 | 0.6 |
| Afghanistan | AFG | 2004 | 2018 | 44 | 3.1 |
| Angola | AGO | 2005 | 2018 | 21 | 1.6 |
| Albania | ALB | 2004 | 2018 | 39 | 2.8 |
| Andorra | AND | 2007 | 2007 | 1 | |
| United Arab Emirates | ARE | 2004 | 2017 | 28 | 2.2 |
| Argentina | ARG | 2004 | 2018 | 19 | 1.4 |
| Armenia | ARM | 2004 | 2018 | 45 | 3.2 |
| Antigua & Barbuda | ATG | 2004 | 2015 | 14 | 1.3 |
| Australia | AUS | 2004 | 2018 | 36 | 2.6 |
| Austria | AUT | 2004 | 2018 | 36 | 2.6 |
| Azerbaijan | AZE | 2004 | 2016 | 19 | 1.6 |
| Burundi | BDI | 2004 | 2015 | 36 | 3.3 |
| Belgium | BEL | 2005 | 2018 | 35 | 2.7 |
| Benin | BEN | 2004 | 2018 | 38 | 2.7 |
| Burkina Faso | BFA | 2004 | 2018 | 42 | 3.0 |
| Bangladesh | BGD | 2004 | 2018 | 36 | 2.6 |
| Bulgaria | BGR | 2004 | 2018 | 30 | 2.1 |
| Bahrain | BHR | 2006 | 2016 | 3 | 0.3 |
| Bahamas | BHS | 2004 | 2018 | 17 | 1.2 |
| Bosnia & Herzegovina | BIH | 2004 | 2018 | 45 | 3.2 |
| Belarus | BLR | 2004 | 2018 | 41 | 2.9 |

Continued on next page

| Country | ccode | First report | Last report | No. of reports | Mean |
|--------------------------|-------|--------------|-------------|----------------|------|
| Belize | BLZ | 2004 | 2018 | 25 | 1.8 |
| Bermuda | BMU | 2005 | 2008 | 4 | 1.3 |
| Bolivia | BOL | 2004 | 2017 | 20 | 1.5 |
| Brazil | BRA | 2012 | 2018 | 27 | 4.5 |
| Barbados | BRB | 2004 | 2018 | 20 | 1.4 |
| Brunei | BRN | 2005 | 2016 | 11 | 1.0 |
| Bhutan | BTN | 2004 | 2018 | 15 | 1.1 |
| Botswana | BWA | 2004 | 2018 | 26 | 1.9 |
| Central African Republic | CAF | 2004 | 2018 | 27 | 1.9 |
| Canada | CAN | 2004 | 2018 | 41 | 2.9 |
| Switzerland | CHE | 2005 | 2018 | 38 | 2.9 |
| Chile | CHL | 2004 | 2018 | 31 | 2.2 |
| China | CHN | 2004 | 2018 | 33 | 2.4 |
| Cote d'Ivoire | CIV | 2007 | 2018 | 35 | 3.2 |
| Cameroon | CMR | 2004 | 2018 | 35 | 2.5 |
| Congo - Kinshasa | COD | 2004 | 2015 | 26 | 2.4 |
| Congo - Brazzaville | COG | 2004 | 2015 | 33 | 3.0 |
| Cook Islands | COK | 2004 | 2004 | 2 | |
| Colombia | COL | 2004 | 2018 | 43 | 3.1 |
| Comoros | COM | 2004 | 2018 | 33 | 2.4 |
| Cape Verde | CPV | 2004 | 2018 | 31 | 2.2 |
| Costa Rica | CRI | 2004 | 2018 | 24 | 1.7 |
| Cayman Islands | CYM | 2005 | 2009 | 4 | 1.0 |
| Cyprus | CYP | 2005 | 2018 | 35 | 2.7 |
| Czechia | CZE | 2004 | 2018 | 30 | 2.1 |
| Germany | DEU | 2004 | 2018 | 45 | 3.2 |

Continued on next page

| Country | ccode | First report | Last report | No. of reports | Mean |
|----------------------------------|-------|--------------|-------------|----------------|------|
| Djibouti | DJI | 2004 | 2017 | 17 | 1.3 |
| Dominica | DMA | 2004 | 2018 | 23 | 1.6 |
| Denmark | DNK | 2004 | 2018 | 37 | 2.6 |
| Dominican Republic | DOM | 2006 | 2018 | 10 | 0.8 |
| Algeria | DZA | 2004 | 2018 | 37 | 2.6 |
| Ecuador | ECU | 2006 | 2016 | 5 | 0.5 |
| Egypt | EGY | 2005 | 2018 | 14 | 1.1 |
| Spain | ESP | 2005 | 2018 | 62 | 4.8 |
| Estonia | EST | 2004 | 2018 | 24 | 1.7 |
| Ethiopia | ETH | 2004 | 2018 | 30 | 2.1 |
| Finland | FIN | 2005 | 2017 | 27 | 2.2 |
| Fiji | FJI | 2004 | 2018 | 8 | 0.6 |
| France | FRA | 2004 | 2018 | 38 | 2.7 |
| Micronesia (Federated States of) | FSM | 2005 | 2017 | 12 | 1.0 |
| Gabon | GAB | 2004 | 2018 | 19 | 1.4 |
| United Kingdom | GBR | 2004 | 2018 | 55 | 3.9 |
| Georgia | GEO | 2004 | 2018 | 48 | 3.4 |
| Guernsey | GGY | 2011 | 2011 | 6 | |
| Ghana | GHA | 2004 | 2018 | 35 | 2.5 |
| Gibraltar | GIB | 2007 | 2007 | 4 | |
| Guinea | GIN | 2004 | 2018 | 36 | 2.6 |
| Gambia | GMB | 2004 | 2018 | 40 | 2.9 |
| Guinea-Bissau | GNB | 2005 | 2018 | 30 | 2.3 |
| Equatorial Guinea | GNQ | 2005 | 2018 | 19 | 1.5 |
| Greece | GRC | 2005 | 2018 | 37 | 2.8 |
| Grenada | GRD | 2004 | 2017 | 19 | 1.5 |

Continued on next page

| Country | ccode | First report | Last report | No. of reports | Mean |
|-------------------|-------|--------------|-------------|----------------|------|
| Guatemala | GTM | 2005 | 2018 | 21 | 1.6 |
| Guyana | GUY | 2004 | 2018 | 10 | 0.7 |
| Honduras | HND | 2004 | 2018 | 21 | 1.5 |
| Croatia | HRV | 2004 | 2018 | 26 | 1.9 |
| Haiti | HTI | 2004 | 2017 | 35 | 2.7 |
| Hungary | HUN | 2004 | 2018 | 37 | 2.6 |
| Indonesia | IDN | 2004 | 2018 | 42 | 3.0 |
| Isle of Man | IMN | 2009 | 2009 | 5 | |
| India | IND | 2004 | 2018 | 31 | 2.2 |
| Ireland | IRL | 2004 | 2018 | 55 | 3.9 |
| Iran | IRN | 2004 | 2018 | 21 | 1.5 |
| Iraq | IRQ | 2004 | 2017 | 23 | 1.8 |
| Iceland | ISL | 2005 | 2018 | 42 | 3.2 |
| Israel | ISR | 2004 | 2018 | 35 | 2.5 |
| Italy | ITA | 2004 | 2017 | 48 | 3.7 |
| Jamaica | JAM | 2004 | 2018 | 33 | 2.4 |
| Jersey | JEY | 2009 | 2009 | 4 | |
| Jordan | JOR | 2004 | 2017 | 26 | 2.0 |
| Japan | JPN | 2004 | 2018 | 45 | 3.2 |
| Kazakhstan | KAZ | 2004 | 2018 | 30 | 2.1 |
| Kenya | KEN | 2004 | 2018 | 34 | 2.4 |
| Kyrgyzstan | KGZ | 2004 | 2017 | 37 | 2.8 |
| Cambodia | KHM | 2004 | 2018 | 23 | 1.6 |
| Kiribati | KIR | 2009 | 2017 | 9 | 1.1 |
| St. Kitts & Nevis | KNA | 2007 | 2017 | 17 | 1.7 |
| South Korea | KOR | 2005 | 2018 | 26 | 2.0 |

Continued on next page

| Country | ccode | First report | Last report | No. of reports | Mean |
|------------------|-------|--------------|-------------|----------------|------|
| Kuwait | KWT | 2004 | 2018 | 30 | 2.1 |
| Laos | LAO | 2004 | 2018 | 24 | 1.7 |
| Lebanon | LBN | 2004 | 2017 | 23 | 1.8 |
| Liberia | LBR | 2004 | 2018 | 40 | 2.9 |
| Libya | LBY | 2005 | 2013 | 12 | 1.5 |
| St. Lucia | LCA | 2004 | 2018 | 14 | 1.0 |
| Liechtenstein | LIE | 2008 | 2018 | 4 | 0.4 |
| Sri Lanka | LKA | 2004 | 2018 | 30 | 2.1 |
| Lesotho | LSO | 2004 | 2018 | 25 | 1.8 |
| Lithuania | LTU | 2005 | 2018 | 30 | 2.3 |
| Luxembourg | LUX | 2004 | 2018 | 28 | 2.0 |
| Latvia | LVA | 2004 | 2018 | 29 | 2.1 |
| Morocco | MAR | 2004 | 2018 | 43 | 3.1 |
| Monaco | MCO | 2008 | 2008 | 1 | |
| Moldova | MDA | 2004 | 2018 | 47 | 3.4 |
| Madagascar | MDG | 2004 | 2018 | 32 | 2.3 |
| Maldives | MDV | 2005 | 2017 | 12 | 1.0 |
| Mexico | MEX | 2004 | 2018 | 71 | 5.1 |
| Marshall Islands | MHL | 2006 | 2018 | 11 | 0.9 |
| Macedonia | MKD | 2004 | 2018 | 30 | 2.1 |
| Mali | MLI | 2004 | 2018 | 53 | 3.8 |
| Malta | MLT | 2005 | 2018 | 15 | 1.2 |
| Myanmar (Burma) | MMR | 2012 | 2018 | 11 | 1.8 |
| Montenegro | MNE | 2008 | 2018 | 22 | 2.2 |
| Mongolia | MNG | 2005 | 2018 | 34 | 2.6 |
| Mozambique | MOZ | 2004 | 2018 | 42 | 3.0 |

Continued on next page

| Country | ccode | First report | Last report | No. of reports | Mean |
|-------------------------|-------|--------------|-------------|----------------|------|
| Mauritania | MRT | 2006 | 2018 | 31 | 2.6 |
| Mauritius | MUS | 2005 | 2018 | 22 | 1.7 |
| Malawi | MWI | 2004 | 2018 | 35 | 2.5 |
| Malaysia | MYS | 2004 | 2018 | 24 | 1.7 |
| Namibia | NAM | 2005 | 2018 | 25 | 1.9 |
| Niger | NER | 2004 | 2018 | 37 | 2.6 |
| Nigeria | NGA | 2004 | 2018 | 36 | 2.6 |
| Nicaragua | NIC | 2004 | 2017 | 19 | 1.5 |
| Netherlands | NLD | 2004 | 2018 | 51 | 3.6 |
| Norway | NOR | 2005 | 2018 | 27 | 2.1 |
| Nepal | NPL | 2004 | 2017 | 19 | 1.5 |
| Nauru | NRU | 2017 | 2017 | 1 | |
| New Zealand | NZL | 2004 | 2018 | 35 | 2.5 |
| Oman | OMN | 2005 | 2015 | 2 | 0.2 |
| Pakistan | PAK | 2004 | 2018 | 35 | 2.5 |
| Panama | PAN | 2006 | 2017 | 21 | 1.9 |
| Peru | PER | 2004 | 2018 | 31 | 2.2 |
| Philippines | PHL | 2004 | 2018 | 35 | 2.5 |
| Palau | PLW | 2004 | 2016 | 13 | 1.1 |
| Papua New Guinea | PNG | 2004 | 2018 | 23 | 1.6 |
| Poland | POL | 2004 | 2018 | 46 | 3.3 |
| Portugal | PRT | 2005 | 2018 | 43 | 3.3 |
| Paraguay | PRY | 2004 | 2017 | 35 | 2.7 |
| Palestinian Territories | PSE | 2018 | 2018 | 1 | |
| Qatar | QAT | 2008 | 2018 | 20 | 2.0 |
| Romania | ROU | 2004 | 2018 | 50 | 3.6 |

Continued on next page

| Country | ccode | First report | Last report | No. of reports | Mean |
|------------------------|-------|--------------|-------------|----------------|------|
| Russia | RUS | 2004 | 2018 | 50 | 3.6 |
| Rwanda | RWA | 2004 | 2018 | 48 | 3.4 |
| Saudi Arabia | SAU | 2006 | 2018 | 21 | 1.8 |
| Serbia and Montenegro | SCG | 2004 | 2006 | 11 | 5.5 |
| Sudan | SDN | 2005 | 2017 | 21 | 1.8 |
| Senegal | SEN | 2005 | 2018 | 40 | 3.1 |
| Singapore | SGP | 2004 | 2018 | 31 | 2.2 |
| Solomon Islands | SLB | 2004 | 2018 | 27 | 1.9 |
| Sierra Leone | SLE | 2004 | 2018 | 36 | 2.6 |
| El Salvador | SLV | 2004 | 2018 | 26 | 1.9 |
| San Marino | SMR | 2004 | 2018 | 19 | 1.4 |
| Somalia | SOM | 2015 | 2018 | 7 | 2.3 |
| Serbia | SRB | 2006 | 2018 | 32 | 2.7 |
| South Sudan | SSD | 2014 | 2017 | 2 | 0.7 |
| S o Tom & Pr ncipe | STP | 2005 | 2018 | 29 | 2.2 |
| Suriname | SUR | 2005 | 2018 | 20 | 1.5 |
| Slovakia | SVK | 2005 | 2018 | 19 | 1.5 |
| Slovenia | SVN | 2004 | 2017 | 26 | 2.0 |
| Sweden | SWE | 2004 | 2017 | 39 | 3.0 |
| Swaziland | SWZ | 2006 | 2017 | 15 | 1.4 |
| Seychelles | SYC | 2004 | 2018 | 25 | 1.8 |
| Syria | SYR | 2005 | 2010 | 8 | 1.6 |
| Turks & Caicos Islands | TCA | 2005 | 2015 | 2 | 0.2 |
| Chad | TCD | 2005 | 2018 | 27 | 2.1 |
| Togo | TGO | 2007 | 2018 | 30 | 2.7 |
| Thailand | THA | 2006 | 2018 | 23 | 1.9 |

Continued on next page

| Country | ccode | First report | Last report | No. of reports | Mean |
|--------------------------|-------|--------------|-------------|----------------|------|
| Tajikistan | TJK | 2004 | 2016 | 27 | 2.2 |
| Timor-Leste | TLS | 2004 | 2016 | 18 | 1.5 |
| Tonga | TON | 2006 | 2018 | 15 | 1.2 |
| Trinidad & Tobago | TTO | 2005 | 2018 | 22 | 1.7 |
| Tunisia | TUN | 2004 | 2018 | 28 | 2.0 |
| Turkey | TUR | 2004 | 2018 | 28 | 2.0 |
| Tuvalu | TUV | 2011 | 2018 | 5 | 0.7 |
| Tanzania | TZA | 2004 | 2018 | 42 | 3.0 |
| Uganda | UGA | 2004 | 2018 | 65 | 4.6 |
| Ukraine | UKR | 2004 | 2017 | 37 | 2.8 |
| Uruguay | URY | 2004 | 2018 | 39 | 2.8 |
| United States | USA | 2004 | 2018 | 54 | 3.9 |
| Uzbekistan | UZB | 2005 | 2018 | 9 | 0.7 |
| St. Vincent & Grenadines | VCT | 2004 | 2017 | 17 | 1.3 |
| British Virgin Islands | VGB | 2010 | 2010 | 4 | |
| Vietnam | VNM | 2004 | 2018 | 25 | 1.8 |
| Vanuatu | VUT | 2005 | 2018 | 9 | 0.7 |
| Samoa | WSM | 2004 | 2018 | 19 | 1.4 |
| Kosovo | XKC | 2010 | 2018 | 21 | 2.6 |
| Yemen | YEM | 2005 | 2014 | 8 | 0.9 |
| South Africa | ZAF | 2004 | 2018 | 40 | 2.9 |
| Zambia | ZMB | 2004 | 2017 | 34 | 2.6 |
| Zimbabwe | ZWE | 2004 | 2017 | 17 | 1.3 |

References

- Apel, M. and M. Grimaldi (2012). The information content of central bank minutes.
- Baker, S. R., N. Bloom, and S. J. Davis (2016). Measuring economic policy uncertainty. *The Quarterly Journal of Economics* 131(4), 1593–1636.
- Beaudry, P. and T. Willems (2018). On the macroeconomic consequences of over-optimism. Technical report, National Bureau of Economic Research.
- Bennani, H. and M. Neuenkirch (2017). The (home) bias of European central bankers: New evidence based on speeches. *Applied Economics* 49(11), 1114–1131.
- Benoit, K. (2018). *Quanteda: Quantitative Analysis of Textual Data*. R package version 1.3.4.
- Blanchard, O. J. and D. Leigh (2013). Growth forecast errors and fiscal multipliers. *American Economic Review* 103(3), 117–20.
- Broome, A. (2015). Back to basics: the great recession and the narrowing of IMF policy advice. *Governance* 28(2), 147–165.
- Choi, H. and H. Varian (2009). Predicting initial claims for unemployment benefits. *Google Inc*, 1–5.
- Dhar, S. (2014). IMF Macroeconomic Policy Advice in the Financial Crisis Aftermath. *IEO Background Paper No. BP/14/07 (Washington: International Monetary Fund)*.
- Dreher, A. (2006). IMF and economic growth: The effects of programs, loans, and compliance with conditionality. *World Development* 34(5), 769–788.
- Dreher, A., J.-E. Sturm, and J. R. Vreeland (2015). Politics and IMF conditionality. *Journal of Conflict Resolution* 59(1), 120–148.
- Edwards, M. S., K. A. Coolidge, and D. A. Preston (2011). Who Reveals? Transparency and the IMF’s Article IV Consultations.
- Gallagher, K. P. and Y. Tian (2017). Regulating capital flows in emerging markets: the IMF and the global financial crisis. *Review of development finance* 7(2), 95–106.
- Gehring, K. and V. Lang (2018). Stigma or Cushion? IMF Programs and Sovereign Creditworthiness.
- Gentzkow, M., B. T. Kelly, and M. Taddy (2017). Text as data.
- Grimmer, J. and B. M. Stewart (2013). Text as data: The promise and pitfalls of automatic content analysis methods for political texts. *Political analysis* 21(3), 267–297.
- IMF (2014). 2014 Triennial Surveillance Review - Staff Background Studies. *International Monetary Fund Policy Papers*.
- Kenny, C. and M. O’Donnell. Do the results match the rhetoric? an examination of world bank gender projects. *CGD Policy Paper 77*.
- Kentikelenis, A. E., T. H. Stubbs, and L. P. King (2016). IMF conditionality and development policy space, 1985–2014. *Review of International Political Economy* 23(4), 543–582.
- Laver, M., K. Benoit, and J. Garry (2003). Extracting policy positions from political texts using words as data. *American Political Science Review* 97(2), 311–331.
- Laver, M. and J. Garry (2000). Estimating policy positions from political texts. *American Journal of Political Science*, 619–634.

- Lombardi, D. and N. Woods (2008). The politics of influence: An analysis of IMF surveillance. *Review of International Political Economy* 15(5), 711–739.
- Lowe, W. (2008). Understanding wordscores. *Political Analysis* 16(4), 356–371.
- Lucca, D. O. and F. Trebbi (2009). Measuring central bank communication: An automated approach with application to FOMC statements. Technical report, National Bureau of Economic Research.
- Mussa, M. and M. Savastano (1999). The IMF approach to economic stabilization. *NBER macroeconomics annual* 14, 79–122.
- Ortiz, I., M. Cummins, J. Capaldo, and K. Karunanethy (2015). The decade of adjustment: A review of austerity trends 2010-2020 in 187 countries.
- Rodrik, D. (2006). Goodbye Washington consensus, hello Washington confusion? A review of the World Bank’s economic growth in the 1990s: Learning from a decade of reform. *Journal of Economic literature* 44(4), 973–987.
- Roy, R. and R. Almeida Ramos (2012). IMF Article IV reports: An analysis of policy recommendations. Technical report, Working Paper, International Policy Centre for Inclusive Growth.
- Sands, P., A. El Turabi, P. A. Saynisch, and V. J. Dzau (2016). Assessment of economic vulnerability to infectious disease crises. *The Lancet* 388(10058), 2443–2448.
- Shin, Y. and M. R. Glennerster (2003). *Is Transparency Good for You, and Can the IMF Help?* Number 3-132. International Monetary Fund.
- Tetlock, P. C. (2007). Giving content to investor sentiment: The role of media in the stock market. *The Journal of finance* 62(3), 1139–1168.